One-Dimensional Variational Ionospheric Retrieval Using Radio Occultation Bending Angles: Part 2 - Validation

Sean Elvidge¹, Sean B. Healy², and Ian Culverwell³

¹University of Birmingham ²ECMWF ³Met Office

June 7, 2023

Abstract

Culverwell et al. (2023) described a new one-dimensional variational (1D-Var) retrieval approach for ionospheric GNSS radio occultation (GNSS-RO) measurements. The approach maps a one-dimensional ionospheric electron density profile, modeled with multiple "Vary-Chap' layers, to bending angle space. This paper improves the computational performance of the the 1D-Var retrieval using an improved background model and validates the approach by comparing with the COSMIC-2 profile retrievals, based on an Abel Transform inversion, and co-located (within 200 km) ionosonde observations using all suitable data from 2020. A three or four layer Vary-Chap in the 1D-Var retrieval shows improved performance compared to COSMIC-2 retrievals in terms of percentage error for the F2 peak parameters (NmF2 and hmF2). Furthermore, skill in retrieval (compared to COSMIC-2 profiles) throughout the bottomside (~90 km to 300 km) has been demonstrated. With a single Vary-Chap layer the performance is similar, but this improves by approximately 40% when using four-layers.

One-Dimensional Variational Ionospheric Retrieval Using Radio Occultation Bending Angles: Part 2 -Validation

S. Elvidge¹, S. B. Healy², and I. D. Culverwell³

¹Space Environment and Radio Engineering (SERENE), University of Birmingham, United Kingdom ²European Centre for Medium-Range Weather Forecasts, Reading, United Kingdom ³Met Office, Exeter, United Kingdom

Key Points:

1

2

3

4

5 6

8

9	• Improved computational performance of the 1D-Var bending angle retrieval is demon-
10	strated with a better background model
11	• Extensive validation of the 1D-Var retrieval approach compared to ionosondes and
12	COSMIC-2 retrievals has been undertaken
13	• The 1D-Var retrieval, using four-layers, is shown to have an 40% reduction in root
14	mean square error compared to COSMIC-2 retrievals

Corresponding author: Sean Elvidge, s.elvidge@bham.ac.uk

15 Abstract

Culverwell et al. (2023) described a new one-dimensional variational (1D-Var) retrieval 16 approach for ionospheric GNSS radio occultation (GNSS-RO) measurements. The ap-17 proach maps a one-dimensional ionospheric electron density profile, modeled with mul-18 tiple "Vary-Chap" layers, to bending angle space. This paper improves the computational 19 performance of the the 1D-Var retrieval using an improved background model and val-20 idates the approach by comparing with the COSMIC-2 profile retrievals, based on an Abel 21 Transform inversion, and co-located (within 200 km) ionosonde observations using all 22 suitable data from 2020. A three or four layer Vary-Chap in the 1D-Var retrieval shows 23 improved performance compared to COSMIC-2 retrievals in terms of percentage error 24 for the F2 peak parameters (NmF2 and hmF2). Furthermore, skill in retrieval (compared 25 to COSMIC-2 profiles) throughout the bottomside (~90 km to 300 km) has been demon-26 strated. With a single Vary-Chap layer the performance is similar, but this improves by 27 approximately 40% when using four-layers. 28

²⁹ Plain Language Summary

Culverwell et al. (2023) presented a new way of estimating ionopsheric electron den-30 sity using the amount of bending experienced by GNSS signals in the upper atmosphere. 31 In this paper, as well as providing extensive validation of the technique, the computa-32 tional performance is improved by using better initial conditions. The validation is done 33 by comparing the newly described approach with that used by the COSMIC-2 satellite 34 constellation using additional data from ground-based sensors known as ionosondes. The 35 newly described technique is found to provide improvements of approximately 40% com-36 pared to a complementary approach using by COSMIC-2. 37

38 1 Introduction

Culverwell et al. (2023) described a one-dimensional variational (1D-Var) ionospheric 39 retrieval that can be applied to the Metop-SG measurement geometry, which will trun-40 cate the ionospheric GNSS-RO measurements around 600 km above the surface. This 41 new, general approach for the ionospheric GNSS-RO retrieval problem, is valid for both 42 the truncated and standard GNSS-RO measurement geometry. A new 1D ionospheric 43 bending angle forward operator was described that computes the L2-L1 bending angle 44 differences as a function of impact parameter, assuming that the ionospheric electron den-45 sity can be modeled with multiple Vary-Chap layers. This approach is close to how GNSS-46 RO data is used in neutral atmosphere applications, such as operational Numerical Weather 47 Prediction (NWP) models (Healy & Thépaut, 2006). 48

Furthermore Culverwell et al. (2023) showed that gradient based minimisation tech-49 niques can be successfully applied to this bending angle retrieval problem, and the non-50 linearity of the Vary-Chap functions was demonstrated to not be problematic. The 1D-51 Var filters out measurement noise and the retrieved electron density solutions are smooth 52 in comparison with the more traditional Abel transform retrieval. The use of bending 53 angles may have implications for how GNSS-RO data is used in ionospheric data assim-54 ilation (DA) systems. Until now, slant TEC values have usually been assimilated in iono-55 spheric DA systems, but these require a correction for the Differential Code Biases (DCB). 56 Since the DCB is usually assumed to be constant during the occultation, it will not im-57 pact the bending angle values derived from the raw phase delays, and so it is not required 58 to assimilate these measurements. 59

The purpose of this paper is to fully validate the 1D-Var ionospheric retrieval using COSMIC-2 bending angles. The 1D-Var and COSMIC-2 retrievals, based on the Abel Transform inversion as implemented by University Corporation for Atmospheric Research (UCAR), hereafter known simply as "COSMIC-2 retrievals", are compared to co-located

	NmF2	Parameters $hmF2$	$H_{0,F2}$	k_{F2}	Number of iterations
Background Analysis	$\begin{array}{c} 2.00 \times 10^{12} \\ 5.66 \times 10^{11} \end{array}$	$\frac{300}{244}$	$50.0 \\ 50.1$	$\begin{array}{c} 0.15\\ 0.14\end{array}$	9
Background Analysis	$\begin{array}{c} 7.00 \times 10^{11} \\ 5.66 \times 10^{11} \end{array}$	$\frac{300}{244}$	$50.0 \\ 50.1$	$\begin{array}{c} 0.15\\ 0.14\end{array}$	11
Background Analysis	$\begin{array}{c} 7.00 \times 10^{11} \\ 5.66 \times 10^{11} \end{array}$	$\begin{array}{c} 250 \\ 244 \end{array}$	$50.0 \\ 49.9$	$\begin{array}{c} 0.15\\ 0.14\end{array}$	6
Background Analysis	$\begin{array}{c} 2.00 \times 10^{11} \\ 5.66 \times 10^{11} \end{array}$	$\frac{300}{244}$	$50.0 \\ 50.1$	$\begin{array}{c} 0.15\\ 0.14\end{array}$	14

 Table 1. Background and analysis parameters for a one-layer Vary-Chap with different background conditions.

(within 200 km) ionosonde observations using all suitable data from 2020, which involves
 a statistical comparison of over 10,000 ionospheric profiles.

2 COSMIC-2

66

The FORMOSAT-7/COSMIC-2 constellation is a constellation of six identical satellites in a low inclination orbit at a nominal altitude of 520-550 km with an inclination of 24° (Anthes & Schreiner, 2019). It follows the successful COSMIC-1 program, which launched in April 2006 (Ho et al., 2020). The COSMIC-2 constellation's primary aim is to observe the atmosphere from a low latitude orbit using RO, supporting operational global weather prediction, tropical weather and climate research, space weather forecasting, and ionospheric research.

⁷⁴ Whilst COSMIC-2 do not provide L1 and L2 bending angles, here we use the deriva-⁷⁵ tive of slantwise TEC S, with respect to the impact parameter $a, \partial S/\partial a$. Culverwell et ⁷⁶ al. (2023) showed that, to a good approximation, this is proportional to the difference ⁷⁷ between the L2 and L1 bending angles, plus an electron density term at the LEO satel-⁷⁸ lite. Here we fit to bending angles with impact heights (impact parameter - radius of cur-⁷⁹ vature) between 100 and 500 km. In contrast COSMIC-2 retrievals use data from 0 up ⁸⁰ to a height of ~700 km at a vertical resolution of 1 Hz (≈ 2 to 3 km).

3 Background Model

In this implementation of a 1D-Var system the *a priori* (background) values are effectively used as a first guess to start the minimisation process rather than a strong constraint on the final 1D-Var solution. For example, using COSMIC-1 measurements provided by the Institut D'Estudis Espacials de Catalunya (IEEC) in Spain using a onelayer Vary-Chap and a variety of background parameters gives the analysis results and the number of iterations until convergence shown in Table 1.

⁸⁸ What is clear from Table 1 is that varying the background model parameters has ⁸⁹ almost no impact on the final analysis (the only difference at all is in the third block with ⁹⁰ an $H_{0,F2}$ value of 49.9 where in all other instances it is 50.1). The main difference be-⁹¹ tween varying the parameters is the impact on the number of iterations required to ar-⁹² rive at the analysis results. This varies from 14 iterations to 6. This suggests that an im-

proved background guess may reduce the number of iterations required for the model 93 to converge. 94

Noting this increased performance when using a more accurate background guess, 95 rather than a fixed set of values, a simple model for these parameters is defined below. 96 Whilst a more complicated background model could be used, this increases the compu-97 tational cost and run times. Since the *a priori* values are not a strong constraint in this 98 system, these cost increases provide no advantages. qq

This peak parameter (NmF2, NmF1, NmE, hmF2, hmF1 and hmE) model is based 100 on Nava et al. (2008); Angling et al. (2018). 101

3.1 NmE

102

1

116

119

121

The peak density of the ionospheric E-region, NmE, is well modelled by a simple 103 function based on the a seasonal relationship with F10.7 and solar zenith angle χ (Leitinger 104 & Kirchengast, 1997; Nava et al., 2008): 105

106
$$NmE = \frac{a_e}{80.616} \sqrt{F_{10.7}} \cos^{0.6}(\chi_{eff}), \tag{1}$$

where a_e is a seasonal term given by (Nava et al., 2008) 107

$$a_e = (1.112 - 0.019s_p)^2, \tag{2}$$

(3)

$$s_p = s \cdot \tanh(0.15\phi),$$

where ϕ is latitude in radians, s = -1 if the month is January, February, November or 110 December, s = 0 if March, April, September or October and s = 1 if May, June, July 111

or August. Finally, χ_{eff} is the effective solar zenith angle given by 112

113
$$\chi_{eff} = \frac{\chi + [90 - 0.24 \exp(20 - 0.2\chi)] \cdot \exp(12(\chi - \chi_0))}{1 + \exp(12(\chi - \chi_0))},$$
(4)

where χ_0 is the zenith angle at night-day transition, 86.23° as given in ITU-R P.2297-114 1 (2019).115

3.2 hmE

Across a range of models including NeQuick and the IRI hmE, the height of the 117 E-region peak density, is usually set to a fixed height. In this model 118

$$hmE = 110 \text{ km},\tag{5}$$

as per the updated NeQuick model from Angling et al. (2018). 120

3.3 NmF2

NmF2, the peak density of the F2-layer, is dependent on a number of factors in-122 cluding geographical longitude, latitude, time, season/day of year and solar activity. A 123 number of approaches can be used to described solar activity, for this simple model the 124 sunspot number (SSN) is used. The NeQuick (Nava et al., 2008) and the Ionospheric 125 Reference Model (IRI; e.g. (Bilitza & Reinisch, 2008)) use the Committee Consultative 126 for Ionospheric Radiowave propagation (CCIR) files to compute NmF2 and M(3000)F2 127 (the ratio of the maximum usable frequency at a distance of 3000 km to the F2 layer crit-128 ical frequency, foF2). The CCIR maps for the F2 parameter consist of 988 coefficients 129 for each month. CCIR provides two sets of coefficients, one for low sunspot numbers and 130 one for high (Haralambous et al., 2021). The coefficients for intermediate levels of so-131 lar activity are determined by linear interpolation (European Commission, 2016). 132

As per (Nava et al. (2008); European Commission (2016) Equations 61–68 and 77) the CCIR coefficients are interpolated for the current date and solar activity levels and then vectors of coefficients for Legendre polynomials are calculated. The NmF2 (electrons/m³) is specified in terms of foF2 (Hz):

$$NmF2 = 0.0124 f o F2^2, (6)$$

137

138

139

141

143

146

where foF2 itself is defined as the sum of nine intermediate terms:

$$foF2 = \sum_{n=1}^{9} foF2_n,$$
 (7)

140 where

$$foF2_1 = \sum_{k=1}^{12} f_k m_k,$$
(8)

142 and

 $foF2_n = \cos^{n-1}(\operatorname{lat}) \sum_{k=1}^{q_n} \left[f_{k_n+2k-1} \cos\left((n-1)\operatorname{lon}\right) + f_{k_n+2k} \sin\left((n-1)\operatorname{lon}\right) \right] m_k$ (9)

144 for n = 2, ..., 9 where

$$q_n = \{12, 12, 9, 5, 2, 1, 1, 1, 1\}, \quad n = 1, \dots, 9$$

$$(10)$$

$$k_n = \{-12, 12, 36, 54, 64, 68, 70, 72, 74\}, \quad n = 1, \dots, 9.$$
(11)

$$m_k$$
 are the "modip" coefficients (fully described in European Commission (2016) equa-
tions 6-16), *lon* and *lat* are the longitude and latitude of interest and the f_j are the 76
CCIR coefficients for the Legendre calculation (European Commission, 2016; Equations
40-50; note that f is written as cf^2 in those equations).

¹⁵¹ **3.4 hmF2**

The calculation of the height of the maximum F2 density, hmF2, is based on the NmF2, NmE and the M(3000)F2 (European Commission, 2016; Equations 69–76). The M(3000)F2 is calculated similarly to NmF2 from Section 3.3 as the sum of seven intermediate terms:

$$M(3000)F2 = \sum_{n=1}^{7} M(3000)F2_n,$$
(12)

157 where

156

158

160

$$M(3000)F2_1 = \sum_{k=1}^7 u_k m_k,$$
(13)

159 and

$$M(3000)F2_n = \cos^{n-1}(\operatorname{lat}) \sum_{k=1}^{r_n} \left[u_{h_n+2k-1} \cos\left((n-1)\operatorname{lon}\right) + u_{h_n+2k} \sin\left((n-1)\operatorname{lon}\right) \right] m_k$$
(14)

161 where

$$r_n = \{8, 6, 3, 2, 1, 1\}$$
 and (15)

$$h_n = \{7, 23, 35, 41, 45, 47\}, \tag{16}$$

for n = 2, ..., 7. u_j are the interpolated CCIR coefficients (European Commission, 2016; Equations 40–51; note that u is written as cm3 in those equations).

Then the hmF2 is defined as (Nava et al., 2008)

$$hmF2 = \frac{1490M}{M + \Delta M} \sqrt{\frac{0.0196M^2 + 1}{1.2967M^2 - 1}} - 176$$
(17)

where M = M(3000)F2 and

$$\Delta M = \frac{0.253}{\rho - 1.215} - 0.012, \tag{18}$$

$$\rho = N \cdot \frac{\exp\left(20\left(N - 1.75\right)\right) + 1.75}{\exp\left(20\left(N - 1.75\right)\right) + 1},\tag{19}$$

$$N = \sqrt{\frac{NmF2}{NmE}}.$$
(20)

172 **3.5 NmF1**

169

171

176

180

The maximum density of the F1-layer, NmF1, is primarily defined in this model as in NeQuick (Nava et al., 2008), in terms of the critical density of the E-Region. Specifically

$$NmF1 = 1.96NmE. (21)$$

177 **3.6 hmF1**

The height of the peak density of the F1-layer, hmF1, is defined here simply as the average of the height of peak density of the E and F2 layer:

$$hmF1 = \frac{hmF2 + hmE}{2} = \frac{hmF2 + 110}{2}.$$
(22)

181

3.7 Using the New Background Model

Table 1 demonstrated the impact of the background model values on the number of iterations needed to converge to a solution with the 1D-Var. Rather than using a fixed set of values for the initial guesses for the Vary-Chap layers but instead using values from the model described in the previous sections has a major impact on the iteration rate and the total number of successful convergences (a successful convergence is defined as a convergence which takes less than 50 iterations), Table 2.

The four-layer Vary-Chap model in Table 2 is an E-, F1-, F2- and topside-layer model. 188 The D-region is excluded as it has very little impact on the overall results (see Culverwell 189 et al. (2023)). It can be seen from Table 2 that using the new background model not only 190 reduces the average number of iterations required for convergence in each case but also 191 improves the overall percentage of observations which do converge. The average reduc-192 tion in the number of observations is 5 iterations and the increased percentage of obser-193 vation convergence means that all test versions of the model have at least a 73.4% con-194 vergence rate (improved from 58.7%). 195

¹⁹⁶ 4 Statistical Performance of the 1D-Var Retrieval

To undertake a rigorous statistical analysis of the 1D-Var retrieval technique two things are needed:

	Background using old parameter conditions	Background using new parameter model			
	Observations which converged	Mean Iterations	Observations which converged	Mean Iterations	
1 layer	98.6%	16	99.3%	11	
2 layers	85.5%	34	92.7%	26	
3 layers	66.7%	28	80.2%	24	
4 layers	58.7%	31	73.4%	28	

Table 2. Percentage of observations which converge within 50 iterations and statistics of the rate of convergence for up to four Vary-Chap layers using the first-guess background conditions described in Table 1 of Culverwell et al. (2023) and the background model decribed here.

199 1. a comparison to independent observations,

2. sufficient retrievals to reduce uncertainty in the analysis.

201

200

4.1 Radio Occultation Observations

The current most abundant, and freely available, source of recent RO observations is from the COSMIC-2 satellites (see Section 2). To perform a rigorous statistical analysis of the 1D-Var retrieval technique all available Level 1B (TEC observations) and Level (ionospheric profiles) products from 2020 have been used (UCAR, 2019). COSMICionospheric retrievals use the calibrated TEC data derived from the L1 and L2 phase differences, which is then calibrated using the method described in CDAAC (Strauss et al., 2020).

In total there are 1,838,920 Level 1B occultations (each containing several hundred TEC observations) and an associated 966,358 Level 2 derived ionospheric profiles available to download in 2020.

4.2 Independent Observations

By stepping through a range of HF frequencies transmitted vertically upwards and measuring the return echoes ionosondes can image the vertical profile of the ionosphere up to the peak density. These observations are both widely assimilated by ionospheric models and also commonly used as reference observations (often incorrectly called 'truth') for comparative studies e.g. (Feltens et al., 2011; Elvidge et al., 2014; Scherliess et al., 2011).

Lin et al. (2020) and Cherniak et al. (2021) have previously validated the COSMICprofiles, in terms of peak density/height, by comparison to ionosonde profiles at eight locations across one month in late 2019 and two months in early 2020 respectively. The Lin et al. (2020) study resulted in the comparison of 135 RO profiles and the Cherniak et al. (2021) study used ~ 2200 profiles.

In this work every ionosonde observation (profiles) from within 200 km of the location of an occultation in 2020 has been used to validate both the COSMIC-2 profile reconstruction as well as the 1D-Var retrieval. Usually ionosonde profiles are "autoscaled" to get the true height information from the observations, but this can, and does, give rise to "autoscaling errors". Ionogram autoscaling is the process of automatically detecting the traces on a graph of time-of-flight against transmitted frequency which can then be used to infer the electron densities. The most commonly used autoscaling software is the

Table 3.	Number	of	COSMIC-2	profiles	analysed
----------	--------	----	----------	----------	----------

Total occultations	1,838,920	100.0%
Number which have associated electron density profiles	966,358	52.5%
Number which have nearby ionosonde observations	$10,\!935$	0.59%
Number which do not contain obvious autoscaling errors	$10,\!612$	0.58%

Automatic Real-Time Ionogram Scaler with True height (ARTIST) (Galkin et al., 2008) which uses a hyperbolic trace fitting method. The best way to overcome these autoscaling errors is to manually scale the ionograms (as was done in Lin et al. (2020)), however that is a time consuming process. Another way to address the problem is by looking at the confidence scores that autoscaling software, such as ARITST, uses to assess the success of the inversion. Only profiles with maximum confidence (100) have been used. That results in 10,935 profiles used in this study.

However, even profiles with the maximum confidence score can still contain errors
(Themens et al., 2022). To address this issue, each ionosonde profile has been examined
and removed from the analysis if they contain obvious errors. Overall this results in 10,612
profiles which are used in the analysis, far more than used in previous studies. The number of observations are summarized in Table 3.

4.3 Analysis Results

Initial analysis of the 1D-Var retrieval, in a similar approach to (Lin et al., 2020), is to look at the performance of the F2 peak in terms of density and altitude compared to the ionosondes. These results can then be compared to the COSMIC-2 electron density profiles. However, before looking at the statistical performance of the two approaches, some quality control (QC) of the peak parameters is required.

An easy first approximation to find the F2 peak parameters is to take the maxi-249 mum density of the electron density profile and associated altitude. Figures 1 and 2 show 250 the resulting probability density plots for the maximum density and the associated al-251 titudes respectively. By comparing the COSMIC-2 height of maximums to the proba-252 bility distribution function of hmF2's using data from the Chilton, UK ionosonde be-253 tween 2000 and 2019 (Figure 2) it is clear that the main bulk of the distributions agree 254 closely. However the second and third peaks centred at 100 km and just greater than 0 255 in the COSMIC-2 data are likely not actually hmF2 values. These likely included both 256 observations of sporadic-E and other errors in the COSMIC-2 profiles (e.g. see exam-257 ples in Figure 3). In this analysis hmF2 values are defined to be between 200 km and 258 500 km; if a value is outside of this range it is excluded from the analysis. 259

260

243

4.4 Statistical Performance

To provide a statistical comparison between the F2 peak parameters the relative percentage error $(\Delta X(\%) = 100(X - X_{ionosonde})/X_{ionosonde})$ of the specification of both NmF2 and hmF2 between these ionosonde observations with both the 1D-Var retrieval with up to four-layers and COSMIC-2 retrieval is shown in Table 4.

From these results it is clear that the 1D-Var retrieval outperforms COSMIC-2 when using three or four layers. There is a 7% relative error in the COSMIC-2 retrieval of NmF2 (which is consistent with the results of Cherniak et al. (2021)) compared with -5.3% and -4.2% when using 1D-Var with three and four layers respectively. However, when only using one-layer the 1D-Var has a -14.8% error in the retrieval of the parameter. For hmF2 the COSMIC-2 error is 2.6% and all versions of the 1D-Var retrieval result in a smaller



Figure 1. Histogram of COSMIC-2 maximum densities.



Figure 2. Probability distribution function of COSMIC-2 maximum densities overlaid with hmF2 data from Chilton, UK ionosonde using data from 2000 to 2019.

Table 4. Statistical performance of retrieving NmF2 and hmF2 from the 1D-Var comparedwith COSMIC-2 relative to ionosondes

	COSMIC-2	1 T	1D-Var	9 T	4 T
		1 Layer	2 Layers	3 Layers	4 Layers
NmF2	7.0%	-14.8%	-10.6%	-5.3%	-4.2%
hmF2	2.6%	1.3%	0.7%	-0.1%	-0.1%



Figure 3. Sample of COSMIC-2 electron density profiles, showing occasional sporadic E-layers and spurious results, COSMIC-2 filenames are above each subplot.

error decreasing from 1.3% to just -0.1%. However it should be noted that errors in hmF2 specification from autoscaled ionosonde observations can be large ($\sim \pm 10\%$) and this should be kept in mind when comparing the parameters.

A more detailed view of how the number of layers improves the F2 specification 274 can be seen by looking at probability density functions of the error in NmF2 specifica-275 tion. Figure 4a shows the probability density function for COSMIC-2 and a single F2 276 layer, overall the distributions are similar, albeit with a slight negative bias in the 1D-277 Var retrieval. Moving around in Figure 4 the 1D-Var retrieval has more layers added to 278 the solution, from one to four. With increasing layers the bias and standard deviation 279 is reduced, with overall the four layers in Figure 4d performing the best, with a signif-280 icantly reduced standard deviation compared to the COSMIC-2 retrieval. 281

282 4.5 B

4.5 Bottomside Profile

Whilst the F2-peak is an important parameter and a good indicator of how well 283 the 1D-Var retrieval is working, one reason for using the Vary-Chap layers is to recon-284 struct the full ionospheric profile. To assess the bottomside (the altitudes region below 285 that of the peak ionospheric density) performance all of the ionospheric profiles (from 286 the ionosonde, COSMIC-2 and the 1D-Var) are interpolated onto the same altitude grid, 287 at 1 km resolution, and the root mean square error (RMSE) with respect to the ionosonde 288 at each altitude is then calculated. The resulting RMSE altitude profile is shown in Fig-289 ure 5. 290

This shows that the 1D-Var retrieval with one-layer (F2) performs very similarly 291 to the COSMIC-2 profiles, and above 225 km there is no statistically significant differ-292 ences in the results, however below 200 km the COSMIC-2 retrieval is clearly superior. 293 A two-layer model (F2+F1) shows an improvement over COSMIC-2 above 200 km, which 294 is consistent with the fact of adding a lower-altitude layer. The four-layer (F2+F1+E+Topside)295 1D-Var retrieval shows an excellent performance throughout the altitude range, with an 296 average improvement over COSMIC-2 by approximately 40%. Only above ~280 km do 297 the COSMIC-2 results show comparable/improved performance relative to the four-layer 298



Figure 4. Panels show comparisons of probability density functions of the error in NmF2 from COSMIC-2 compared to the 1D-Var solution with (a) an F2 Layer, (b) F2+F1 layers, (c) F2+F1+E layers and (d) F2+F1+E+Topside layers. Vertical dashed line is marked at 0.



Figure 5. Root mean square error (RMSE) altitude profiles for COSMIC-2 (red) and the 1D-Var retrieval with only 1 Layer (F2; blue), 2 Layers (F2+F1; orange) and 4 Layers (F2+F1+E+Topside; green) compared to ionosonde observations.

model, although the number of data points at these altitudes is relatively small and the
 results are not statistically distinguishable.

5 Discussion and Conclusions

The statistical performance of the 1D-Var retrieval technique has been thoroughly evaluated in the current work, comparing it to independent observations and leveraging a substantial set of retrievals to reduce uncertainty in the analysis. The method was applied to the Level 1B and Level 2 data obtained from COSMIC-2 satellites and validated using ionosonde observations, constituting a rigorous and expansive analytical process.

From the analysis, it was clear that the 1D-Var retrieval technique demonstrates 308 a robust performance, outperforming COSMIC-2 retrievals, particularly when employ-309 ing three or four layers. The relative percentage errors in the specification of both NmF2 310 and hmF2 were found to be lower for the 1D-Var retrieval. This demonstrates the sig-311 nificant potential of the 1D-Var technique in the field of ionospheric profile reconstruc-312 tion. However, it's worth noting that the 1D-Var retrieval's performance was significantly 313 lower when only one layer was used, indicating the importance of multi-layer modeling 314 for improved accuracy. This is an important finding and suggests that future work should 315 focus on refining and utilizing multi-layer models to improve the retrieval of ionospheric 316 parameters. 317

Furthermore, the comprehensive analysis of the full ionospheric profiles showed that the four-layer 1D-Var retrieval exhibited excellent performance throughout the altitude range. Notably, there was an average improvement over COSMIC-2 by approximately 40%, showcasing the promising potential of this technique. Although the COSMIC-2 results showed comparable performance to the four-layer model above 280 km, the number of data points at these altitudes was relatively small, and the results weren't statistically distinguishable. This suggests that further research should be dedicated to improving the modeling and retrieval at these higher altitudes.

In conclusion, the 1D-Var retrieval technique, particularly when using three or four layers, offers a significant advancement in ionospheric profile reconstruction. While there are still areas for improvement, particularly in the retrieval at higher altitudes, the findings of this work provide a strong foundation for further research in this field.

330 6 Open Research

COSMIC-1 data is available from https://data.cosmic.ucar.edu/gnss-ro/cosmic1/repro2013/. The processed files from the IEEC may be requested from Dr Hernández-Pajares. COSMIC-2 data is available from https://data.cosmic.ucar.edu/gnss-ro/cosmic2/nrt/ and the ionosonde observations were retrieved from The National Centers for Environmental Information (NCEI) https://www.ngdc.noaa.gov/stp/iono/ionogram.html.

The MODIP and CCIR files required to reconstruct the *a priori* model described in this work are available from the International Telecommunications Union (https://www.itu.int/md/R07-WP3L-C-0094/en) or by contacting Bruno Nava (https://t-ict4d.ictp.it/nequick2/sourcecode).

The background model described in this paper is coded in Python 3.x, as is the 1DVar retrieval code that uses it. An officially supported Fortran95 implementation of the latter has been part of the Radio Occultation Processing Package (ROPP) since version 11.0 (released January 2022). ROPP is maintained, developed and supported by EUMET-SAT, through the Radio Occultation Meteorology Satellite Applications Facility (ROM SAF), and freely available to download from its website (ROM SAF, 2023).

346 Acknowledgments

We thank Dr Haixia Lyu and Dr Hernández-Pajares for providing the COSMIC-1 data used here. Dr Riccardo Notarpietro and Dr David Themens are thanked for useful discussions during the early stages of this work.

This work was conducted as part of the Visiting Scientist program of the Radio Occultation Meteorology Satellite Applications Facility (ROM SAF) which is a decentralized operational radio occultation processing center under EUMETSAT. S. Elvidge was a ROM SAF Visiting Scientist for this project, and S. Healy and I. Culverwell are respectively current and former members of the ROM SAF.

355 References

- Angling, M. J., Elvidge, S., & Healy, S. B. (2018, April). Improved model for correcting the ionospheric impact on bending angle in radio occultation measurements. Atmospheric measurement techniques, 11(4), 2213-2224. Retrieved 2019-02-21, from https://www.atmos-meas-tech.net/11/2213/2018/ doi: 10.5194/amt-11-2213-2018
- Anthes, R., & Schreiner, W. (2019). Six new satellites watch the atmosphere over earth's equator. *Eos*, 100. doi: https://doi.org/10.1029/2019EO131779
- Bilitza, D., & Reinisch, B. (2008, August). International Reference Ionosphere
 2007: Improvements and new parameters. Advances in Space Research, 42(4),
 599-609. Retrieved 2019-02-21, from https://linkinghub.elsevier.com/
 retrieve/pii/S0273117708000288 doi: 10.1016/j.asr.2007.07.048

367	Cherniak, I., Zakharenkova, I., Braun, J., Wu, Q., Pedatella, N., Schreiner, W.,
368	Hunt, D. (2021). Accuracy assessment of the quiet-time ionospheric F2 peak
369	parameters as derived from COSMIC-2 multi-GNSS radio occultation mea-
370	surements. Journal of Space Weather and Space Climate, 11, 18. Retrieved
371	2021-09-11, from https://www.swsc-journal.org/articles/swsc/abs/
372	2021/01/swsc200070/swsc200070.html (Publisher: EDP Sciences) doi:
373	10.1051/swsc/2020080
374	Culverwell, I. D., Healy, S. B., & Elvidge, S. (2023). One-dimensional variational
375	ionospheric retrieval using radio occultation bending angles: Part 1 - theory.
376	Journal of Geophysical Research: Space Physics, $X(Y)$.
377	Elvidge, S., Angling, M. J., & Nava, B. (2014, September). On the use of modified
378	Taylor diagrams to compare ionospheric assimilation models. Radio Science,
379	49(9), 737-745. Retrieved 2019-02-21, from http://doi.wiley.com/10.1002/
380	2014RS005435 doi: 10.1002/2014RS005435
381	European Commission. (2016). Ionospheric correction algorithm for galileo single
382	frequency users (Tech. Rep.). Retrieved from https://www.gsc-europa.eu/
383	sites/default/files/sites/all/files/Galileo_Ionospheric_Model.pdf
384	Feltens J Angling M J Jackson-Booth N Jakowski N Hoque M Hernández-
385	Pajares, M., Zandbergen, R. (2011, December). Comparative testing of
386	four ionospheric models driven with GPS measurements. <i>Radio Science</i> , 46(6).
387	Retrieved 2019-02-21. from http://doi.wilev.com/10.1029/2010RS004584
388	doi: 10.1029/2010RS004584
380	Galkin I Khmyrov G Kozlov A Reinisch B W Huang X & Paznukhov V
300	(2008) The ABTIST 5 Radio Sounding and Plasma Physics AIP Proceedings
301	$\pm 97\%$
303	Haralambous H Leontiou T Petrou V Kumar Singh A Charalambides M
392	Lithoxopoulos N & Agisilaou A (2021) Adjusting ccir maps to improve
393	local behaviour of ionospheric models Atmosphere 12(6) Betrieved from
394	https://www.mdpi_com/2073-4433/12/6/691_doi: $10.3390/atmos12060691$
395	Healy S B & Thépaut I N (2006) Assimilation experiments with champ gros
390	radio occultation measurements. <i>Quarterly Journal of the Royal Meteorological</i>
397	Society 139 doi: https://doi.org/10.1256/ai.04.182
398	Ho S p Anthea R A Ao C O Healy S Heranyi A Hunt D Zong Z
399	(2020 July) The COSMIC/FORMOSAT 3 Radio Occultation Mission after
400	12 Vorse: Accomplishments, Remaining Challenges, and Potential Impacts of
401	COSMIC 2 Bullatin of the American Meteorological Society 101(7) E1107-
402	F1136 Batriavad 2023 04 28 from https://iournals.amotsoc.org/uiou/
403	iournals/hams/101/7/hamsD180200_yml (Publisher: American Meteoro
404	logical Society Section: Bulletin of the American Meteorological Society) doi:
405	10 1175/BAMS-D-18-0200 1
400	ITU P P 2207 1 (2010) Floatron density models and data for transion or horiz radio
407	monagation (Toch Bon) Betrieved from https://www.itu.int/dms.pub/itu
408	-r/onb/ren/B-BED-D 2207-1-2019-DDE-E ndf
409	Latinger P. & Kirchengest C. (1007) Fact to Use Clobal and Pagional Ione
410	spheric Models A Benert on Approaches Used in Craz
411	Hung 20 220 242 doi: https://doi.org/10.1007/BE02225504
412	Hing, 52, 525-542. doi: https://doi.org/10.1001/DF05525504
413	LIII, U I., LIII, U. U I., LIU, J I., Rajesii, P. K., Matsuo, I., Oliou, M I., Vah W. H. (2020 October) The Farly Decults and Validation of
414	EORMOSAT 7/COSMIC 2 Space Weather Droducts. Clobal Ionegration Cross
415	ification and No Aided Abal Floatron Donaity Drofie Lowrond of Combusical
416	Research, Space Drawing 105(10) 020201102020 Detwiewood 2021 00 01 from
417	http://agupubg.onlinelibrary.uiley.com/doi/10_1020/20201402029
418	(Publisher: John Wiley & Song Itd) doi: 10.1020/2020/202028
419	Nava B. Coisson P. & Radicalla S. (2008) A new version of the network ione
420	sphere electron density model Journal of Atmospheric and Solar Terrestrial
441	

422	Physics, 70(15), 1856-1862.
423	ROM SAF. (2023). ROPP Software Deliverable. Retrieved from https://www
424	.romsaf.org/ropp/
425	Scherliess, L., Thompson, D. C., & Schunk, R. W. (2011). Data assimilation models:
426	A new tool for ionospheric science and applications. In W. Liu & M. Fujimoto
427	(Eds.), The Dynamics Magnetosphere (pp. 329–339). Springer, Berlin.
428	Strauss, P., Schreiner, W., & Santiago, J. (2020). FORMOSAT-7/COSMIC-2
429	TGRS Space Weather Provisional Data Release 1 (Tech. Rep.). Retrieved
430	from https://data.cosmic.ucar.edu/gnss-ro/cosmic2/provisional/
431	spaceWeather/F7C2_SpWx_Provisional_Data_Release_1.pdf
432	Themens, D. R., Reid, B., & Elvidge, S. (2022). Artist ionogram autoscaling confi-
433	dence scores: Best practices. Radio Science Letters - Under Review.
434	UCAR. (2019). COSMIC-2 Data Products (Tech. Rep.). Retrieved from
435	https://data.ucar.edu/en/dataset/cosmic-2-data-products doi:
436	10.5065/t353-c093

436